



Global Warming: Experts' Opinions versus Scientific Forecasts¹

by

Kesten C. Green

**Business and Economic Forecasting Unit,
Monash University**

and

J. Scott Armstrong

**The Wharton School,
University of Pennsylvania**

NCPA Policy Report No. 308

February 2008

ISBN #1-56808-183-9

Web site: www.ncpa.org/pub/st/st308

National Center for Policy Analysis

12770 Coit Rd., Suite 800

Dallas, Texas 75251

(972) 386-6272

Executive Summary

In 2007, the Intergovernmental Panel on Climate Change (IPCC) issued its *Fourth Assessment Report*. The report included predictions of big increases in average world temperatures by 2100, resulting in an increasingly rapid loss of the world's glaciers and ice caps, a dramatic global sea level rise that would threaten low-lying coastal areas, the spread of tropical diseases, and severe drought and floods.

These dire predictions are not, however, the result of scientific forecasting; rather, they are the opinions of experts. Expert opinion on climate change has often been wrong. For instance, a search of headlines in the *New York Times* found the following:

Sept. 18, 1924	MacMillan Reports Signs of New Ice Age
March 27, 1933	America in Longest Warm Spell Since 1776
May 21, 1974	Scientists Ponder Why World's Climate Is Changing: A Major Cooling Widely Considered to be Inevitable

Problems with Computer Models. Climate scientists now use computer models, but there is no evidence that modeling improves the accuracy of predictions. For example, according to the models, the Earth should be warmer than actual measurements show it to be. Furthermore:

- The General Circulation Models (GCMs) that are used failed to predict recent global average temperatures as accurately as fitting a simple curve to the historical data and extending it into the future.
- The models forecast greater warming at higher altitudes in the tropics, whereas the data show the greatest warming has occurred at lower altitudes and at the poles.
- Furthermore, individual models have produced widely different forecasts from the same initial conditions, and minor changes in assumptions can produce forecasts of global cooling.

Skepticism Among the Scientists. Thus it is not surprising that international surveys of climate scientists from 27 countries in 1996 and 2003 found growing skepticism over the accuracy of climate models. Of more than 1,060 respondents, only 35 percent agreed with the statement, "Climate models can accurately predict future climates," whereas 47 percent disagreed.

Violations of Forecasting Principles. Forty internationally-known experts on forecasting methods and 123 expert reviewers codified evidence from research on forecasting into 140 principles. The empirically-validated principles are available in the *Principles of Forecasting* handbook and at forecastingprinciples.com. These principles were designed to be applicable to making forecasts about

diverse physical, social and economic phenomena, from weather to consumer sales, from the spread of nonnative species to investment strategy, and from decisions in war to egg-hatching rates. They were applied to predicting the 2004 U.S. presidential election outcome and provided the most accurate forecast of the two-party vote split of any published forecast, and did so well ahead of election day (see polyvote.com).

The authors of this study used these forecasting principles to audit the IPCC report. They found that:

- Out of the 140 forecasting principles, 127 principles are relevant to the procedures used to arrive at the climate projections in the IPCC report.
- Of these 127, the methods described in the report violated 60 principles.
- An additional 12 forecasting principles appear to be violated, and there is insufficient information in the report to assess the use of 38.

As a result of these violations of forecasting principles, the forecasts in the IPCC report are invalid. Specifically:

The Data Are Unreliable. Temperature data is highly variable over time and space. Local proxy data of uncertain accuracy (such as ice cores and tree rings) must be used to infer past global temperatures. Even over the period during which thermometer data have been available, readings are not evenly spread across the globe and are often subject to local warming from increasing urbanization. As a consequence, the trend over time can be rising, falling or stable depending on the data sample chosen.

The Forecasting Models Are Unreliable. Complex forecasting methods are only accurate when there is little uncertainty about the data and the situation (in this case: how the climate system works), and causal variables can be forecast accurately. These conditions do not apply to climate forecasting. For example, a simple model that projected the effects of Pacific Ocean currents (El Niño-Southern Oscillation) by extrapolating past data into the future made more accurate three-month forecasts than 11 complex models. Every model performed poorly when forecasting further ahead.

The Forecasters Themselves Are Unreliable. Political considerations influence all stages of the IPCC process. For example, chapter by chapter drafts of the *Fourth Assessment Report "Summary for Policymakers"* were released months in advance of the full report, and the final version of the report was expressly written to reflect the language negotiated by political appointees to the IPCC. The conclusion of the audit is that there is no scientific forecast supporting the widespread belief in dangerous human-caused "global warming." In fact, it has yet to be demonstrated that long-term forecasting of climate is possible.

Introduction

More than 20 years ago, scientists began to express concern that human activities — primarily tropical deforestation and the burning of fossil fuels for energy — threaten to cause a rapid warming of the Earth by adding carbon dioxide (CO₂) to the atmosphere. Recognizing the problems that global warming might cause, in 1998 the World Meteorological Organization (WMO) and the United Nations Environment Program (UNEP) established the Intergovernmental Panel on Climate Change (IPCC).² The purpose of the IPCC was to provide a comprehensive, objective, scientific, technical and socio-economic assessment of the current understanding of human-induced climate change, its potential impacts and options for adaptation and mitigation.

In 2007, the IPCC issued its *Fourth Assessment Report*. The *Assessment* in fact consists of three reports and a “synthesis” report. The first part was titled “The Physical Science Basis” and was authored by the IPCC’s Working Group One (WG1), a panel of experts on climate science, modeling and history. This paper focuses on the first report.³ It included predictions of dramatic increases in average world temperatures by 2100, which might in turn cause such serious environmental harms as: a global sea level rise that would threaten low-lying coastal areas, the spread of tropical diseases, an increasingly rapid loss of the world’s glaciers and ice caps, and a worsening of drought and flooding events across broad regions.⁴

Although the IPCC’s 1,056-page report makes these dire predictions, nowhere does it refer to empirically-validated forecasting methods, despite the fact these are conveniently available in books and articles and on Web sites. These evidence-based forecasting principles have been validated through experiment and testing and comparison to actual outcomes. The evidence shows that adherence to the principles increases forecast accuracy. This paper uses these scientific forecasting principles to ask: Are the IPCC’s forecasts a good basis for developing public policy? The answer is “no.”

Three elements are necessary for governments to make rational policy responses to climate change: Scientists must accurately predict (1) global temperature, (2) the effects of any temperature changes and (3) the effects of feasible alternative policy responses. At any step in this process, the failure to obtain a valid forecast would render forecasts at the next step in the process meaningless. This study focuses primarily, but not exclusively, on the first of the three forecasts required: obtaining long-term forecasts of global temperature. It finds that due to the unscientific method by which these forecasts were obtained, they cannot be relied upon. [See the sidebar, “Three Forecasts Required for Climate Change Policies.”]

“Climate change policies must be based on accurate, scientific forecasts.”

Three Forecasts Required for Climate Change Policies

Before they can determine the best policies to deal with the climate of the future, policymakers must select an appropriate statistic to use to represent the changing climate. By convention, the statistic is the averaged global temperature as measured with thermometers at ground stations throughout the world. However, in practice this is a far from satisfactory metric. For instance:

- Stations for taking temperature readings are absent from large areas of the Earth's surface.
- Temperature measurements are biased due to the urban-heat-island effect, whereby weather stations located originally on the edge of town became surrounded by buildings, factories, roads, cars, car parks and airport tarmacs; temperature measures from such stations don't accurately represent the temperature for the larger surrounding area.
- The locations and number of measuring stations have changed, which calls into question the consistency of measurements over time and the comparability of past measurements to present ones.¹

However, assuming the problems with obtaining accurate current temperature measurements is resolved, the first step is to obtain forecasts and prediction intervals for the long-term mean global temperature (say 20 years or longer).

If accurate forecasts of mean global temperature can be obtained *and* the changes are substantial, the second step is to obtain forecasts of the effects of temperature changes on humans and other living things. Concerns about changes in global mean temperature assume the earth is currently at the optimal temperature and that variations over years (unlike variations *within* days and years) are undesirable, but a proper assessment requires comprehensive projections of both the costs and the benefits of changes from the current global average temperature.²

If valid forecasts of the effects of the temperature changes on the health of living things and on the health and wealth of humans can be obtained *and* substantial harmful effects are forecast, the third step is to calculate the costs and the benefits of proposed alternative policy responses. A policy proposal should only be implemented *if* valid and reliable forecasts of the effects of implementing the policy can be obtained *and* the forecasts show net benefits.

¹ C. Essex, R. McKittrick and B. Andresen, "Does a global temperature exist?" *Journal of Non-Equilibrium Thermodynamics*, Vol. 32, No. 1, February 2007, pages 1-27. Working paper available at <http://www.uoguelph.ca/~rmckitri/research/globaltemp/globaltemp.html>. Access verified December 10, 2007.

² For example, policy responses to Rachel Carson's *Silent Spring* should have been based in part on forecasts of the number of people who might die from malaria if DDT use were reduced. Instead, the use of DDT was banned solely on the basis of predicted harm to bird species. Consideration of the one million human deaths and between 300 and 500 million cases of malaria annually might have led to a policy other than outright proscription. United Nations International Children's Emergency Fund (UNICEF), "Malaria," undated. Available at http://www.unicef.org/health/index_malaria.html. Access verified December 10, 2007.

Scientific Forecasting versus Opinion

Scientific forecasting methods should be used to project climate change. The methodologies used should be those shown empirically to be relevant to the particular types of problems involved in climate forecasting. However, the evidence shows that the IPCC forecasts are instead based on opinions.

Many public-policy decisions are based on unaided expert judgments. The experts may have access to empirical studies and other information, but they often make predictions, or judgmental forecasts, without the aid of scientific forecasting principles. Research on persuasion has shown that people have substantial faith in the value of such forecasts, and this faith increases when experts agree with one another. However, the opinions of experts are an invalid basis for public policy.

Judgmental Forecasts. Comparative empirical studies have routinely concluded that judgmental forecasting by experts is the least accurate of the methods available to make forecasts.⁵ For example, Professor Phil Tetlock, of the University of California at Berkeley, conducted a major study in which he recruited 284 participants whose professions included, “commenting or offering advice on political and economic trends.”⁶ He asked these experts to forecast the probability that various events would or would not occur, picking areas (geographic and substantive) within and outside their areas of expertise. By 2003, he had accumulated over 82,000 forecasts. The experts barely outperformed nonexperts and neither group did well against simple forecasting rules that extrapolate from the past to predict the future.⁷

Examples of expert climate forecasts that turned out to be completely wrong are also easy to find. [See the sidebar on “Climate Forecasts Based on Expert Opinion.”]

Computer Modeling versus Scientific Forecasting. Over the past few decades, the methodology used in climate forecasting has shifted so that expert opinions are informed by computer models. Advocates of complex climate models claim that they are based on well-established laws of physics. But there is clearly much more to the models than physical laws, otherwise the models would all produce the same output, which they do not, and there would be no need for confidence estimates for model forecasts, which there certainly is. Climate models are, in effect, mathematical ways for experts to express their opinions.⁸

There is no empirical evidence that presenting opinions in mathematical terms rather than in words improves the accuracy of forecasts. In the 1800s, Thomas Malthus forecast mass starvation. Expressing his opinions in a mathematical model, he predicted that the food supply would increase arithmetically while the human population would grow at a geometric rate and go hungry. Mathematical models have not become much more accurate since.

“The opinions of experts are wrong as often as the opinions of nonexperts.”

Climate Forecasts Based on Expert Opinion

Contradictory contemporary climate forecasts by experts are easy to find. Some have proved to be wildly inaccurate. For example, in a speech on Earth Day, April 22, 1970, University of California at Davis ecologist Kenneth Watt predicted, “If present trends continue, the world will be about four degrees colder in 1990, but eleven degrees colder in the year 2000. This is about twice what it would take to put us into an ice age.”

A few years later, in the mid-1970s, political debate raged about whether the global climate was changing. The United States’ National Defense University (NDU) addressed this issue in a study that provided a chart that showed the mean annual temperature rising from 1870 to early 1940 then dropping sharply up to 1970.¹ The study concluded that while a slight increase in temperature might occur, uncertainty was so high that it was most likely that the experience of “the next twenty years will be similar to that of the past,” and the effects of any change would be negligible. This conclusion was based primarily on a survey that asked experts their opinion regarding future temperature changes and weighted the 19 replies. Clearly, this was a judgmental forecast by scientists, based on their informed opinion, not a scientific forecast. It is generally agreed that temperatures have risen since 1970, rather than following the declining 1940 to 1970 trend. However, the NDU study was influential. It was discussed in *The Global 2000 Report to the President* (Carter) and at the World Climate Conference in Geneva in 1979.

Experts’ forecasts of climate changes have long been newsworthy and a cause of worry for people. For instance, a search of headlines in the *New York Times* found the following:²

Sept. 18, 1924	MacMillan Reports Signs of New Ice Age
March 27, 1933	America in Longest Warm Spell Since 1776
May 21, 1974	Scientists Ponder Why World’s Climate is Changing: A Major Cooling Widely Considered to be Inevitable

The forecasts behind these headlines were made with a high degree of confidence.³

¹ National Defense University, *Climate Change to the Year 2000* (Washington, D.C.: National Defense University, 1978).

² R.W. Anderson and D. Gainor, “Fire and Ice: Journalists have warned of climate change for 100 years, but can’t decide whether we face an ice age or warming,” Business and Media Institute, May 17, 2006. Available at <http://www.businessandmedia.org/specialreports/2006/fireandice/FireandIce.pdf>. Access verified December 10, 2007.

³ An earlier review of empirical research on this problem led to the “Seer-sucker theory,” which can be stated as “No matter how much evidence exists that seers do not exist, seers will find suckers.” J. Scott Armstrong, “The Seer-sucker theory: The value of experts in forecasting,” *Technology Review*, Vol. 83, June-July 1980, pages 16-24. Available at <http://129.3.20.41/eps/get/papers/0412/0412009.pdf>. Access verified December 10, 2007.

Two international surveys of climate scientists from 27 countries, in 1996 and 2003, show increasing skepticism over the accuracy of climate models.⁹ Of more than 1,060 respondents, only 35 percent agreed with the statement, "Climate models can accurately predict future climates," whereas 47 percent disagreed. [See Figure I.]¹⁰

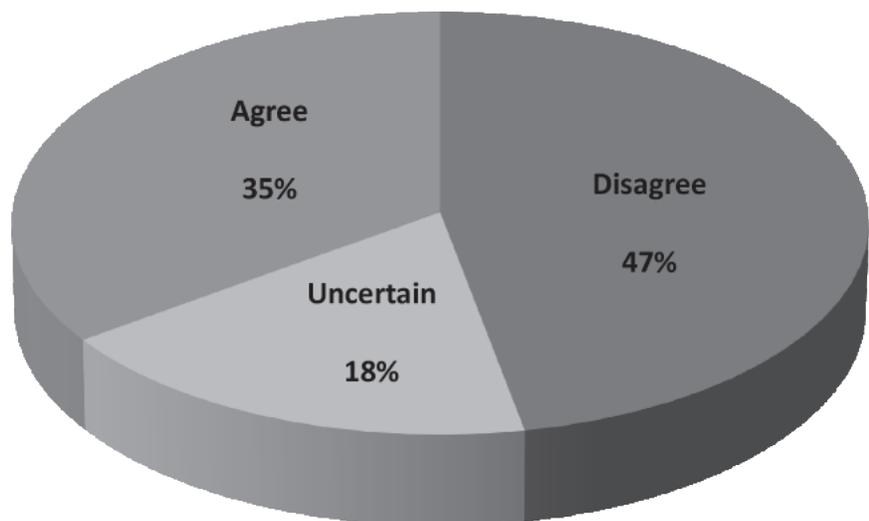
Problems with Climate Models. Researchers who have examined long-term climate forecasts have concluded they are based on nothing more than scientists' opinions expressed in complex mathematical terms, without valid evidence to support the chosen approach.¹¹

For example, when computer simulations project future global mean temperatures with twice the current level of atmospheric CO₂, they assume that the temperature forecast is as accurate as a computer simulation of present temperatures with current levels of CO₂.¹² Yet it has never been demonstrated that temperature forecasts are as accurate as simulations of current conditions based on actual temperature data. Indeed, there are even serious questions surrounding model simulations of current temperature with current CO₂ levels. According to the models, the earth should be warmer than actual measurements show it to be, which is why modelers adjust their findings to fit the data.

The models do not represent the real world sufficiently well to be relied upon for forecasting.¹³ As physicist Freeman Dyson concluded, climate mod-

FIGURE I

Climate Models Can Accurately Predict Future Climates



"Computer modeling does not improve the accuracy of opinion."

Source: Joseph Bast and James M. Taylor, *Scientific Consensus on Global Warming* (Chicago: Heartland Institute, 2007).

“Climate models based on historical data don’t accurately predict current temperatures.”

els “do a very good job of describing the fluid motions of the atmosphere and the oceans,” but “they do a very poor job of describing the clouds, the dust, the chemistry and the biology of fields and farms and forests.”¹⁴

The climate models’ simulations do not correspond to past or present temperatures, rainfall patterns, tropical cyclones or plant responses. Professor Bob Carter, of the Marine Geophysical Laboratory at James Cook University in Australia, examined evidence on the predictive ability of the general circulation models (GCMs) used by the IPCC scientists. He found that while the models included some basic principles of physics, scientists had to make a number of “educated guesses” because knowledge about the physical processes of the earth’s climate is incomplete.¹⁵ Thus, in practice:

- The GCMs failed to predict recent global average temperatures as accurately as fitting a simple curve to the historical data and extending it into the future.
- The models forecast greater warming at higher altitudes in the tropics, when the greatest warming has occurred at lower altitudes and at the poles.
- Furthermore, individual models have produced widely different forecasts from the same initial conditions, and minor changes in their assumptions can produce forecasts of global cooling.¹⁶

When models predict global cooling, the forecasts are rejected by modelers as “outliers” or “obviously wrong.” This suggests that when the models are averaged together to create consensus estimates of temperature change, the results are biased due to the omission of models that show cooling.¹⁷

Researchers have found serious deficiencies in the GCMs on which the IPCC based its previous *Third Assessment Report*. For example, David Belamy and Jack Barrett found:¹⁸

- 1) The models produced very different cloud distributions and none came close to the actual distribution of clouds, which is important because the type and distribution of clouds can either enhance or reduce the Earth’s temperature. Some clouds tend to trap more of the sun’s radiant heat, while others reflect the sun’s rays away from the Earth’s surface, mitigating the effect of increased greenhouse gases.
- 2) Assumptions about the amount of radiation from the Sun absorbed by the atmosphere and the Earth’s surface varied considerably from model to model, yielding widely varying temperature forecasts. This is important because absorption of heat energy from sunlight is the primary mechanism of global warming.
- 3) The models did not accurately represent the known effects of CO₂, much less the uncertain possible feedbacks that reduce or enhance

those effects, and as a result their climate change forecasts cannot be relied upon.

The review by Bellamy and Barrett concluded: "The climate system is a highly complex system and, to date, no computer models are sufficiently accurate for their [IPCC] predictions of future climate to be relied upon."¹⁹ And since climate model forecasts for periods of up to five years have proven to be inaccurate, the review concluded there is little basis upon which to make accurate projections for 50 to 100 years.²⁰

Referring to the GCMs used for the IPCC forecasts, a lead author of Chapter 3 of the *Fourth Assessment Report* wrote that "... the science is not done because we do not have reliable or regional predictions of climate."²¹

Other agencies' attempts to forecast for shorter periods and smaller geographical areas have also been unsuccessful. For instance, annual forecasts by New Zealand's National Institute of Water and Atmospheric Research (NIWA) are no more accurate than chance.²² NIWA's low success rate is comparable to other forecasting groups worldwide, according to New Zealand climatologist Jim Renwick, a member of the IPCC Working Group I and a coauthor of the *Fourth Assessment Report*. (Renwick also serves on the World Meteorological Organization Commission for Climatology Expert Team on Seasonal Forecasting.) He concludes that current GCMs are unable to predict future global climate any better than chance.²³

"Climate models based on recent data can't accurately predict temperatures five years in the future."

An Audit of the IPCC Report

To what extent have those who have made climate forecasts used scientifically tested forecasting procedures? Using forecasting audit software, the authors of this study independently assessed the extent to which IPCC procedures conformed to or violated forecasting principles.

Climate Forecasters' Use of the Scientific Literature on Forecasting Methods.²⁴ There is little use of forecasting principles in environmental research generally; apparently they are not used at all in climate research.²⁵ An Internet search found no Web sites or papers on climate change that referenced forecasting methodology.²⁶ Neither the IPCC Report's Chapter 8, "Climate models and their evaluation," nor any of the 788 referenced works therein, refer to forecasting methodology or established forecasting principles.²⁷ The same was true of Chapter 9, "Understanding and attributing climate change," and its 535 references. A survey of climate scientists (described below) did not yield references to any relevant papers on forecasting.²⁸

Forecasting principles have been derived from all known empirical evidence on estimating the as yet unknown. The principles are therefore scientific. Evidence comes from all disciplines that have produced relevant evidence and the principles are applicable to all forecasting problems — from

Principles of Forecasting

Research on forecasting has been conducted since the 1930s. Empirical studies that compare methods and sources of evidence have been conducted to determine which forecasts are the most accurate. Researchers began to summarize, gather, collate and systematize these findings in the 1970s and 1980s.¹ In the mid-1990s, the Forecasting Principles Project was established with the objective of summarizing all useful knowledge about forecasting. The knowledge was codified as evidence-based principles. Evidence-based principles tell which methods to use under what conditions, based on the evidence to date. The strongest form of evidence comes from empirical studies that have compared the predictive ability of different forecasting methods.

The Forecasting Principles Project led to the *Principles of Forecasting* handbook: the work of 40 internationally-known experts on forecasting methods and 123 reviewers who were also leading experts on forecasting methods.² The forecasting principles are available on a site maintained and updated by the International Institute of Forecasters.³ For example, these forecasting principles were applied to the problem of predicting the outcome of the 2004 U.S. presidential election, and yielded predictions of a win for President Bush early and consistently, and predicted the vote more accurately than any other forecaster.

Following are examples of the principles that apply to long-term forecasts for complex situations where the causal factors are subject to uncertainty (as with climate).

Principle: Unaided judgmental forecasts by experts have no value. “Unaided” means evidence-based forecasting principles were not used. Such forecasts are (at best) educated guesses. This principle applies whether the opinions are expressed in words, spreadsheets or mathematical models. It also applies regardless of how much scientific evidence is possessed by the experts. Among the reasons for this are:

- a) Complexity: People cannot assess complex relationships through unaided observations.
- b) Coincidence: People confuse correlation with causation.
- c) Feedback: People making judgmental predictions typically do not receive the explicit feedback they could use to improve their forecasting.
- d) Bias: People have difficulty in obtaining or using evidence that contradicts their initial beliefs. This problem is especially serious for people who view themselves as experts.

Principle: Agreement among experts is weakly related to accuracy. This is especially true when the experts communicate with one another and when they work together to solve problems, as is the case with the IPCC process.

Principle: Complex models involving nonlinear relationships and interactions between variables are less accurate when there is uncertainty because errors multiply. For instance, the widely publicized 1972 forecasts by the Club of Rome of ecological and economic collapse proudly proclaimed, “in our model, about 100,000 relationships are stored in the computer.”⁴ However, the model made wildly inaccurate predictions, such as widespread famine and resource depletion by the year 2000.

Complex models tend to fit random variations in historical data well, but as a consequence of their complexity, their forecasts are less accurate than simpler models and result in misleading conclusions about outcomes. Furthermore, when complex models are developed there are many opportunities for errors, and the complexity makes errors more difficult to find. For example, long-term energy forecasts based on computer models failed to predict the ability of the U.S. economy to increase energy efficiency in response to oil embargos in the 1970s.⁵

Principle: Given even modest uncertainty, prediction intervals are enormous. Prediction intervals (ranges outside which outcomes are unlikely to fall) expand rapidly as time horizons increase. The longer the time from the date of the prediction to the date of the predicted event or result, the greater the likelihood of an unanticipated result (one falling outside the expected range of outcomes). This is true, for example, even when trying to forecast something as straightforward as automobile sales for General Motors over the next five years.

Principle: When there is uncertainty, forecasts should be conservative. Uncertainty arises when data contain measurement errors, when the series are unstable, when knowledge about the direction of relationships is uncertain, and when a forecast depends upon forecasts of related (causal) variables. For example, forecasts of no change were found to be more accurate than trend forecasts for annual sales when there was substantial uncertainty in the trend lines.⁶

¹ J. Scott Armstrong, *Long-Range Forecasting: From Crystal Ball to Computer* (New York, N.Y.: Wiley-Interscience, 1985).

² J. Scott Armstrong, *Principles of Forecasting* (Norwell, Mass.: Kluwer Academic Press, 2001).

³ See forecastingprinciples.com, a site sponsored by the International Institute of Forecasters. A summary of the principles, currently numbering 140, is provided as a checklist in the Forecasting Audit software available on the site. The site is often updated in order to incorporate new evidence on forecasting as it comes to hand. J. Scott Armstrong, “Findings from evidence-based forecasting: Methods for reducing forecast error,” *International Journal of Forecasting*, Vol. 22, 2006, pages 583-598.

⁴ W. Ascher, *Forecasting: An Appraisal for Policy Makers and Planners* (Baltimore, Md.: Johns Hopkins University Press, 2005).

⁵ P. P. Craig, A. Gadgil and J. G. Koomey, “What Can History Teach Us? A Retrospective Examination of Long-Term Energy Forecasts for the United States,” *Annual Review of Energy and the Environment*, Vol. 27, 2002, pages 83-118.

⁶ This principle also implies that forecasts should revert to long-term trends when such trends have been firmly established, do not waver, and there are no firm reasons to suggest they will change. Finally, trends should be damped toward no change as the forecast horizon increases. See S. P. Schnaars and R. J. Bavuso, “Extrapolation models on very short-term forecasts,” *Journal of Business Research*, Vol. 14, 1986, pages 27-36.

weather to company sales, from the spread of non-native species to investment strategy, and from war fighting to egg hatching rates. [See the sidebar on Principles of Forecasting.]

Forecasting Audit Results. Of the 140 forecasting principles, the audit found that 127 forecasting principles are relevant to the procedures used to arrive at the climate projections in the IPCC report. Of these 127, the methods described in the report definitely violate 60 principles, 12 appear to be violated and there is insufficient information to assess the use of 38.

For example, “Make sure forecasts are independent of politics (Principle 1.3),” is one of the 60 principles the IPCC process clearly violated. David Henderson, a former Head of Economics and Statistics at the OECD, has given a detailed account of how political considerations influence all stages of the IPCC process.²⁹ For example, the “Summary for Policymakers” that accompanies each of the IPCC’s assessment reports is released with much media and public fanfare. The summary is written in negotiation with the explicit input of legislators, policymakers and/or diplomatic appointees. Most recently, chapter by chapter drafts of the *Fourth Assessment Report’s “Summary for Policymakers”* were released months in advance of the final version of the full report, with the directive that the final version of the chapters in the report be expressly written to reflect the language negotiated by the lead authors with the participating political appointees to the IPCC.

“Official climate reports do not use scientific forecasting principles.”

Three Basic Forecasting Principles Violated by the Fourth Assessment Report

Some principles are so important that any forecasting process that does not adhere to them cannot produce valid forecasts. The following are three such principles, all of which are based on strong empirical evidence, and all of which were violated by the forecasting procedures described in Chapter 8 of the IPCC report.

Principle: Consider whether the events or series can be forecasted.

This principle refers to whether a forecast would likely be more accurate than assuming things will not change. Predicting no change is to use a naïve forecasting method. It is appropriate to use a naïve method when knowledge is poor and uncertainty is high, as with climate.³⁰ There is even controversy among climate scientists over something as basic as the current trend with respect to temperature. Temperature data is highly variable and cyclical, and proxy data (such as ice cores and tree rings) must be used to infer temperatures more than a few decades past. Whether a trend over time is determined to be rising, falling or stable depends largely on the beginning and endpoint chosen.³¹

“Temperature data is often incomplete and unreliable.”

Global climate is complex and scientific evidence on key relationships is weak or absent. For example, the IPCC holds that CO₂ plays a significant causal role in determining global temperature, but a number of studies have presented evidence that causation generally works in the opposite direction (increases in atmospheric levels of CO₂ are a result of higher temperatures) and that CO₂ variation plays at most a minor role in climate change.³²

Measurements of key variables such as local temperatures and a representative global temperature are contentious and subject to revision. In the case of modern measurements they must be adjusted for the changing distribution of weather stations and such complicating factors as the urban-heat-island effect. The interpretation of proxy data for ancient temperatures is often speculative.³³ Finally, it is difficult to forecast the causal variables — for example CO₂ and cloudiness.³⁴

Although the authors of Chapter 8 claim that the forecasts of global mean temperature are well-founded, their language is imprecise and relies heavily on such words as “generally,” “reasonably well,” “widely,” and “relatively.” These terms indicate that the IPCC forecasts are uncertain, and the chapter makes many explicit references to uncertainty.³⁵

In discussing temperature modeling, the authors wrote, “The extent to which these systematic model errors affect a model’s response to external perturbations is unknown, but may be significant,” and, “The diurnal temperature range... is generally too small in the models, in many regions by as much as 50 percent,” and “It is not yet known why models generally underestimate the diurnal temperature range.”³⁶

Given the high uncertainty regarding climate, the appropriate naïve method for this situation would be the “no change” model since prior evidence suggests that attempts to improve upon the naïve model often increase forecast error. To reverse this conclusion, one would have to produce validated evidence in favor of alternative methods. Chapter 8 of the IPCC report does not provide this evidence. If long-term forecasting of climate is possible, it has yet to be demonstrated.

Principle: Keep forecasting methods simple. IPCC chapters and related papers leave the impression that climate forecasters believe that complex models are necessary for forecasting climate and that forecast accuracy will increase with model complexity. Complex methods involve such things as the use of a large number of variables in forecasting models and complex interactions between variables. Complex forecasting methods are only accurate when there is little uncertainty about relationships now and in the future, where the data are subject to little error, and where the causal variables can be accurately forecast. These conditions do not apply to climate forecasting. Thus, simple methods are recommended.

“The models are complex and produce conflicting results.”

The use of complex models when uncertainty is high conflicts with the evidence from forecasting research.³⁷ For example, scientists Halmar Halide and Peter Ridd compared predictions of El Niño-Southern Oscillation events from a simple extrapolation of the time series with those from other researchers’ complex models.³⁸ Some of the complex models were dynamic causal models incorporating laws of physics. In other words, they were similar to those upon which the IPCC authors depended. The simple model made more accurate three-month predictions than all 11 of the complex models. Every model performed poorly when forecasting further ahead.

Using complex methods prevents understanding of how forecasts were derived, makes criticism difficult, and error detection unlikely.

Principle: Do not use fit to develop the model. It is unclear to what extent the models described in the IPCC report are either based on, or have been tested against, sound empirical data.³⁹ However, some statements were made about the ability of the models to fit historical data, after tweaking their parameters. Extensive research has shown that the ability of a model to reproduce historical data has little relationship to forecast accuracy.⁴⁰ Fit can be improved by making a model more complex. The typical consequence of increasing complexity to improve fit, however, is to decrease the accuracy of forecasts.

Other Audit Results. The audit also found 12 “apparent violations.” These principles are areas where the authors had concerns over the coding or did not agree that the procedures clearly violated the principle. Finally, for many of the relevant principles, there was insufficient information to make ratings. Some of these principles might be surprising to those who are not familiar with forecasting research — for example: “Use all important variables (Principle 10.2).” Others are principles that any scientific paper should be expected to address, such as: “Use objective tests of assumptions.” And others are especially important to climate forecasting, such as: “Limit subjective adjustments of quantitative forecasts.”⁴¹

The number of violations of forecasting principles found by the audit confirms the survey finding described earlier that the IPCC authors were unaware of forecasting principles.⁴² Had they been aware of the principles, it would have been incumbent on them to present evidence to justify their departures from them. They did not do so. Because the forecasting processes examined in Chapter 8 overlook scientific evidence on forecasting, the IPCC forecasts of climate change are not scientific.

Climate change forecasters should use the readily available Forecasting Audit program to ensure that they are using appropriate forecasting procedures. Outside evaluators should also be encouraged to conduct audits. These reports should be made available to both study sponsors and the public by posting on an open Web site such as publicpolicyforecasting.com.

Conclusion

Three elements are necessary for governments to make rational policies in response to climate change: Scientists must accurately predict (1) global temperature changes, (2) the effects of any temperature changes and (3) the effects of feasible alternative policy responses. To justify policy changes, governments need scientific forecasts for all three forecasting problems and they need those forecasts to show net benefits flowing from proposed policies. If governments implement policy changes without such justification, they are likely to cause harm.

This paper has shown that failure occurs with the first forecasting problem: predicting temperature over the long term. Specifically, no scientific forecast supports the widespread belief in dangerous human-caused “global warming.” Climate is complex and there is much uncertainty about causal relationships and data. Prior research on forecasting suggests that in such situations a naïve (no change) forecast would be superior to current predictions. Note that recommending the naïve forecast does not mean that climate will not change. It means that current knowledge about climate is insufficient to make useful long-term forecasts about climate. Policy proposals should be assessed on that basis.

Many policies have been proposed in association with claims of global warming. This paper does not purport to comment on specific policy proposals, but it should be noted that some policies may be valid regardless of future climate changes. To assess this, it would be necessary to directly forecast costs and benefits that assume: (1) that climate does not change and (2) that climate changes in a variety of ways.

The evidence shows that those forecasting long-term climate change have limited or no apparent knowledge of evidence-based forecasting methods; therefore, similar conclusions apply to the second two elements of the forecasting problem. Public policy makers owe it to the people who would be affected by their policies to base them on scientific forecasts. Advocates of policy changes have a similar obligation. Hopefully, climate scientists with diverse views will begin to embrace forecasting principles and will collaborate with forecasting experts in order to provide policy makers with scientific climate forecasts.

“There is no scientific forecast supporting the belief in human-caused global warming.”

NOTE: Nothing written here should be construed as necessarily reflecting the views of the National Center for Policy Analysis or as an attempt to aid or hinder the passage of any bill before Congress.

Notes

¹ This paper is based substantially on Kesten C. Green and J. Scott Armstrong, “Global Warming: Forecasts by Scientists versus Scientific Forecasts,” *Energy and Environment*, Vol. 18, Nos. 7 & 8, 2007, pages 997-1,021.

² Intergovernmental Panel on Climate Change, “About IPCC,” undated. Available at <http://www.ipcc.ch/about/index.htm>. Access verified December 10, 2007.

³ Intergovernmental Panel on Climate Change, *Climate Change 2007: The Physical Science Basis* (New York: Cambridge University Press, 2007). Available at <http://ipcc-wg1.ucar.edu/wg1/wg1-report.html>. Access verified December 10, 2007.

⁴ Many observers have claimed the IPCC presents scenarios or projections rather than forecasts. See K. E. Trenberth, “Predictions of Climate,” *Climate Feedback: The Climate Change Blog*, Nature.com, June 4, 2007. Available at http://blogs.nature.com/climatefeedback/2007/06/predictions_of_climate.html. Access verified December 10, 2007. A forecast or prediction is an estimate of actual values in a future time period (for time series) or for another situation (for cross-sectional data). The terms “scenario” and “projection” appear to be used in the IPCC’s *Fourth Assessment Report* to indicate that it provides “conditional forecasts.” As it happens, in Chapter 8, “Climate models and their evaluation,” the report uses the word “forecast” and its derivatives 37 times, and “predict” and its derivatives 90 times. Furthermore, the majority (31 of the 51) of scientists who responded to a survey for this study agreed that the IPCC report is the most credible source of *forecasts* (not “scenarios” or “projections”) of global average temperature.

⁵ W. Ascher, *Forecasting: An Appraisal for Policy Makers and Planners* (Baltimore, Md.: Johns Hopkins University Press, 1978).

⁶ P.E. Tetlock, *Expert Political Judgment: How Good Is It? How Can We Know?* (Princeton, N.J.: Princeton University Press, 2005).

⁷ J. Scott Armstrong, “Extrapolation for time-series and cross-sectional data,” in J. S. Armstrong, ed., *Principles of Forecasting* (Norwell, Mass.: Kluwer Academic Press, 2001).

⁸ Reid Bryson, the most-cited climatologist in academic studies worldwide, said a model is “*nothing more* than a formal statement of how the modeler believes that the part of the world of his concern actually works.” Reid A. Bryson, “Environment, Environmentalists, and Global Change: A Skeptic’s Evaluation,” *New Literary History*, Vol. 24, No. 4, 1993, pages 783-795.

⁹ Joseph Bast and James M. Taylor, *Scientific Consensus on Global Warming* (Chicago, Ill.: Heartland Institute, 2007). Available at <http://downloads.heartland.org/20861.pdf>. Access verified December 10, 2007. (It includes the responses to all questions in the 1996 and 2003 surveys by Dennis Bray and Hans von Storch as an appendix.)

¹⁰ Polls show that the general public is also skeptical of the ability of climate models to accurately portray the future: 40 percent agreed that “climate change was too complex and uncertain for scientists to make useful forecasts,” while 38 percent disagreed. Paul Eccleston, “Public ‘in denial’ about climate change,” *Telegraph.co.uk*, March 7, 2007. Available at <http://www.telegraph.co.uk/earth/main.jhtml?xml=/earth/2007/07/03/eawarm103.xml>. Access verified December 10, 2007.

¹¹ Orrin H. Pilkey and Linda Pilkey-Jarvis, *Useless Arithmetic: Why Environmental Scientists Can’t Predict the Future* (New York, N.Y.: Columbia University Press, 2007).

¹² Bryson, “Environment, Environmentalists, and Global Change.”

¹³ Numerous studies have concluded that the use of climate models for forecasting has failed. See, for example, Robert C. Balling, “Observational surface temperature records versus model predictions,” in P. J. Michaels, ed., *Shattered Consensus: The True State of Global Warming* (Lanham, Md.: Rowman & Littlefield, 2005), pages 50-71; John Christy, “Temperature Changes in the Bulk Atmosphere: Beyond the IPCC,” in Michaels, *Shattered Consensus*, pages 72-105; Oliver W. Frauenfeld, “Predictive Skill of the El Nino-Southern Oscillation and Related Atmospheric Teleconnections,” in Michaels, *Shattered Consensus*, pages 149-182; Eric S. Posmentier and Willie Soon, “Limitations of Computer Predictions of the Effects of Carbon Dioxide on Global Temperature,” in Michaels, *Shattered Consensus*, pages 241-281.

¹⁴ Freeman J. Dyson, “Heretical Thoughts About Science and Society,” *Edge: The Third Culture*, August 8, 2007. Available at http://www.edge.org/3rd_culture/dysonf07/dysonf07_index.html. Access verified December 10, 2007.

¹⁵ R. M. Carter, “The myth of dangerous human-caused climate change,” The Australasian Institute of Mining and Metallurgy New Leaders Conference, Brisbane, Queensland, May 3, 2007, pages 61-74. Available at http://members.iinet.net.au/~glrmc/new_page_1.htm. Access verified December 10, 2007.

- ¹⁶ C. Essex and R. McKittrick, *Taken by Storm: The Troubled Science, Policy & Politics of Global Warming* (Toronto, Ca.: Key Porter Books, 2002).
- ¹⁷ D.A. Stainforth et al., "Uncertainty in predictions of the climate response to rising levels of greenhouse gases," *Nature*, Vol. 433, 2005, pages 403-406.
- ¹⁸ David Bellamy and Jack Barrett, "Climate stability: an inconvenient proof," *Proceedings of the Institution of Civil Engineers – Civil Engineering*, Vol. 160, May 2007, pages 66-72.
- ¹⁹ *Ibid.*, page 72.
- ²⁰ Noted by former Colorado State Climatologist Roger Pielke Sr. See "Interview By Marcel Crok Of Roger A. Pielke Sr.," April 30, 2007. Available at <http://tinyurl.com/2wpk29>. Access verified December 10, 2007.
- ²¹ Trenberth, "Predictions of climate."
- ²² M. Taylor, "An evaluation of NIWA's climate predictions for May 2002 to April 2007," Climate Science Coalition, 2007. Available at <http://nzclimatescience.net/images/PDFs/climateupdateevaluationtext.pdf>. Data available at <http://nzclimatescience.net/images/PDFs/climateupdateevaluationcalc.xls.pdf>.
- ²³ New Zealand Climate Science Coalition, "World climate predictors right only half the time," *Scoop Independent News* (New Zealand), media release, June 7, 2007. Available at <http://www.scoop.co.nz/stories/SC0706/S00026.htm>. Access verified December 10, 2007.
- ²⁴ This study is the only *comprehensive* review of climate forecasting efforts to date, though there have been more limited reviews of some aspects of the IPCC's forecasting principles and assessments of the actual accuracy of the forecasts made. For example, the National Defense University (NDU) forecasting process (described above) was criticized in an audit by T. R. Stewart and M. H. Glantz for lacking awareness of proper forecasting methods. The audit was hampered because the organizers of the NDU study said that the raw data had been destroyed. Judging from a Google Scholar search for citations, climate forecasters have paid little attention to that paper. T. R. Stewart and M. H. Glantz, "Expert judgment and climate forecasting: A methodological critique of 'Climate Change to the Year 2000,'" *Climate Change*, Vol. 7, 1985, pages 159-183.
- ²⁵ Bryson, "Environment, Environmentalists, and Global Change," pages 783-795.
- ²⁶ The best selling textbook on forecasting methods was referenced twice in a search for the term "global warming," but neither citation related to the prediction of global mean temperatures. S. Makridakis, S. C. Wheelwright and R. J. Hyndman, *Forecasting: Methods and Applications*, 3rd ed. (Hoboken, N.J.: John Wiley, 1998).
- ²⁷ As discussed in the sidebar, the majority of scientists responding to our inquiries concerning what climate forecasts were most credible and how they were determined referenced Chapters 8 through 10 of the IPCC WG1 2007 report, *Climate Change 2007*. Available at <http://ipcc-wg1.ucar.edu/wg1/wg1-report.html>. Access verified December 10, 2007.
- ²⁸ Including Chapter 10 of the IPCC Report, Gerald A. Meehl et al., "Global Climate Projections," in *Climate Change 2007*, pages 747-846. Available at http://ipcc-wg1.ucar.edu/wg1/Report/AR4WG1_Print_Ch10.pdf. Access verified December 10, 2007.
- ²⁹ David R. Henderson, "Governments and Climate Change Issues: The Case for Rethinking," *World Economics*, Vol. 8, No.2, 2007, pages 183-228. Available at http://forecastingprinciples.com/Public_Policy/Henderson2007paper.pdf. Access verified December 10, 2007.
- ³⁰ Freeman Dyson wrote that "The real world is muddy and messy and full of things that we do not yet understand." Dyson, "Heretical Thoughts About Science and Society."
- ³¹ Accurate direct measurements of tropospheric global average temperature have only been available since 1979, and they show no evidence for greenhouse warming. Surface thermometer data, though flawed, show temperature stasis since 1998. Carter, "The myth of dangerous human-caused climate change."
- ³² H. Le Treut et al., "Historical Overview of Climate Change," in IPCC, *Climate Change 2007*. Available at http://ipcc-wg1.ucar.edu/wg1/Report/AR4WG1_Print_Ch01.pdf. Access verified December 10, 2007. See also, Willie Soon, "Implications of the secondary role of carbon dioxide and methane forcing in climate change: Past, present and future," *Physical Geography*, July 4, 2007. Available at <http://arxiv.org/ftp/arxiv/papers/0707/0707.1276.pdf>. Access verified December 10, 2007.
- ³³ Carter, "The myth of dangerous human-caused climate change."
- ³⁴ Even with perfect knowledge of emissions, uncertainties in the representation of atmospheric and oceanic processes by climate models limit the accuracy of any estimate of the climate response. Natural variability, generated both internally and from external forcings such as changes in solar output and explosive volcanic eruptions, also contributes to the uncertainty

in climate forecasts. P. A. Stott and J. A. Kettleborough, “Origins and estimates of uncertainty in predictions of twenty-first century temperature rise,” *Nature*, Vol. 416, 2002, pages 723-726.

³⁵ For example, the phrases “It is not yet possible to determine which estimates of the climate change cloud feedbacks are the most reliable,” and “Despite advances since the TAR (Third Assessment Report), substantial uncertainty remains in the magnitude of cryospheric feedbacks within AOGCMs.” (Cryospheric feedbacks are responses of frozen areas of the earth — the world’s Arctic and Antarctic regions, and glaciers — to climate changes. AOGCMs are general circulation models of the ocean/atmosphere make-up and interaction.) The following words and phrases appear at least once in the Chapter: unknown, uncertain, unclear, not clear, disagreement, not fully understood, appears, not well observed, variability, variety, unresolved, not resolved, and poorly understood. David A. Randall et al., “Climate Models and their Evaluation,” in IPCC, *Climate Change 2007*, page 593. Available at http://ipcc-wg1.ucar.edu/wg1/Report/AR4WG1_Print_Ch08.pdf. Access verified December 10, 2007.

³⁶ *Ibid.*, pages 608-609.

³⁷ P. G. Allen and R. Fildes, “Econometric Forecasting,” in Armstrong, *Principles of Forecasting*; J. Scott Armstrong, *Long-Range Forecasting: From Crystal Ball to Computer* (New York, N.Y.: Wiley-Interscience, 1985); G. T. Duncan, W. L. Gorr and J. Szczypula, “Forecasting Analogous Time Series,” in Armstrong, *Principles of Forecasting*; D. Wittink and T. Bergstuen, “Forecasting with Conjoint Analysis,” in Armstrong, *Principles of Forecasting*.

³⁸ H. Halide and P. Ridd, “Complicated ENSO models do not significantly outperform very simple ENSO models,” *International Journal of Climatology*, 2007, in press.

³⁹ G. C. Hegerl et al., “Understanding and Attributing Climate Change,” in IPCC, *Climate Change 2007*. Available at http://ipcc-wg1.ucar.edu/wg1/Report/AR4WG1_Print_Ch09.pdf. Access verified December 10, 2007.

⁴⁰ Armstrong, *Principles of Forecasting*.

⁴¹ As noted, the ratings in a forecasting audit involve some judgment. As a check on the auditors’ judgments, the detailed ratings were sent to 240 of the scientists and policymakers surveyed earlier for information on the use of forecasting principles. They were asked to review the paper and to identify any inaccuracies. They were also invited to make their own ratings for publication at publicpolicyforecasting.com. As of this writing, none have done so.

⁴² The authors’ ratings of the processes used to generate the forecasts presented in the IPCC report are provided on the Public Policy Forecasting Special Interest Group Page at forecastingprinciples.com. These ratings were posted in late June 2007, when the authors’ paper was presented at the International Symposium on Forecasting in New York.

⁴³ David A. Randall et al., “Climate Models and their Evaluation.”

⁴⁴ Because reliability is an issue with rating tasks, the authors sent out a general request for experts and several colleagues to use the Forecasting Audit Software to conduct their own audits. At this writing, none have done so.

⁴⁵ The final ratings are fully disclosed in the Special Interest Group section of the forecastingprinciples.com site that is devoted to Public Policy (publicpolicyforecasting.com) under Global Warming.

Appendix A

Survey to Identify the Most Credible Long-Term Forecasts of Global Temperature

The authors surveyed scientists involved in long-term climate forecasting and policymakers. The primary goal was to identify the most important forecasts and how they were made. In particular, the analysts wished to know if the most widely accepted forecasts of global average temperature were based on expert opinions or on scientific forecasting methods. This review of current climate forecasting methods along with an extensive online database search found that many scientists are unaware of evidence-based findings related to forecasting methods.

A questionnaire was sent to experts who had expressed diverse opinions on global warming. Lists of experts were generated through the identification of key people and asking them to identify others. (The lists are provided in Appendix Table A.) Most (70 percent) of the 240 experts on the lists were IPCC reviewers and authors.

The questionnaire asked the experts to provide references for what they regarded as the most credible source of long-term forecasts of mean global temperatures. Designed for simplicity, the survey asked which forecasts people regarded as the most credible and how those forecasts were derived, “In your opinion, which scientific article is the source of the most credible forecasts of global average temperatures over the rest of this century?”

Useful responses were received from 51 of those surveyed, 42 of whom provided references to what they regarded as credible sources of long-term forecasts of mean global temperatures. Interestingly, eight respondents provided claims that no credible forecasts exist. Of the 42 expert respondents who appeared to support the hypothesis of manmade global warming (by providing references to forecasts), 30 referenced the IPCC report. A list of the papers suggested by respondents is provided at publicpolicyforecasting.com in the “Global Warming” section.

Based on the survey replies, it was clear that the IPCC’s Working Group 1 Report contained the forecasts that are viewed as most credible by the majority of the climate forecasting community. These forecasts are contained in Chapter 10 of the Report and the models that are used to forecast climate are assessed in Chapter 8, “Climate Models and Their Evaluation.”⁴³ Because Chapter 8 provided the most useful information on the forecasting process used by the IPCC to derive forecasts of mean global temperatures, the authors audited that chapter.

Appendix Table A

People Surveyed for this Study

(* indicates a relevant response)

IPCC Working Group 1

Myles Allen, Richard Alley, Ian Allison, Peter Ambenje, Vincenzo Artale, Paulo Artaxo, Alphonsus Baede, Roger Barry, Terje Berntsen, Richard A. Betts, Nathaniel L. Bindoff, Roxana Bojariu, Sandrine Bony, Kansri Boonpragob, Pascale Braconnot, Guy Brasseur, Keith Briffa, Aristita Busuioc, Jorge Carrasco, Anny Cazenave, Anthony Chen*, Amnat Chidthaisong, Jens Hesselbjerg Christensen, Philippe Ciais*, William Collins, Robert Colman*, Peter Cox, Ulrich Cubasch, Pedro Leite Da Silva Dias, Kenneth L. Denman, Robert Dickinson, Yihui Ding, Jean-Claude Duplessy, David Easterling, David W. Fahey, Thierry Fichefet*, Gregory Flato, Piers M. de F. Forster*, Pierre Friedlingstein, Congbin Fu, Yoshiyuki Fuji, John Fyfe, Xuejie Gao, Amadou Thierno Gaye*, Nathan Gillett*, Filippo Giorgi, Jonathan Gregory*, David Griggs, Sergey Gulev, Kimio Hanawa, Didier Hauglustaine, James Haywood, Gabriele Hegerl*, Martin Heimann*, Christoph Heinze, Isaac Held*, Bruce Hewitson, Elisabeth Holland, Brian Hoskins, Daniel Jacob, Bubu Pateh Jallow, Eystein Jansen*, Philip Jones, Richard Jones, Fortunat Joos, Jean Jouzel, Tom Karl, David Karoly*, Georg Kaser, Vladimir Kattsov, Akio Kitoh, Albert Klein Tank, Reto Knutti, Toshio Koike, Rupa Kumar Kolli, Won-Tae Kwon, Laurent Labeyrie, René Laprise, Corrine Le Quéré, Hervé Le Treut, Judith Lean, Peter Lemke, Sydney Levitus, Ulrike Lohmann, David C. Lowe, Yong Luo, Victor Magaña Rueda, Elisa Manzini, Jose Antonio Marengo, Maria Martelo, Valérie Masson-Delmotte, Taroh Matsuno, Cecilie Mauritzen, Bryant Mcavane, Linda Mearns, Gerald Meehl, Claudio Guillermo Menendez, John Mitchell, Abdalah Mokssit, Mario Molina, Philip Mote*, James Murphy, Gunnar Myhre, Teruyuki Nakajima, John Nganga, Neville Nicholls, Akira Noda, Yukihiro Nojiri, Laban Ogallo, Daniel Olago, Bette Otto-Bliesner, Jonathan Overpeck*, Govind Ballabh Pant, David Parker, Wm. Richard Peltier, Joyce Penner*, Thomas Peterson*, Andrew Pitman, Serge Planton, Michael Prather*, Ronald Prinn, Graciela Raga, Fatemeh Rahimzadeh, Stefan Rahmstorf, Jouni Räisänen, Srikanthan (S.) Ramachandran, Veerabhadran Ramanathan, Venkatachalam Ramaswamy, Rengaswamy Ramesh, David Randall*, Sarah Raper, Dominique Raynaud, Jiawen Ren, James A. Renwick, David Rind, Annette Rinke, Matilde M. Rusticucci, Abdoulaye Sarr, Michael Schulz*, Jagadish Shukla, C. K. Shum, Robert H. Socolow*, Brian Soden, Olga Solomina*, Richard Somerville*, Jayaraman Srinivasan, Thomas Stocker, Peter A. Stott*, Ron

Stouffer, Akimasa Sumi, Lynne D. Talley, Karl E. Taylor*, Kevin Trenberth*, Alakkat S. Unnikrishnan, Rob Van Dorland, Ricardo Villalba, Ian G. Watterson*, Andrew Weaver*, Penny Whetton, Jurgen Willebrand, Steven C. Wofsy, Richard A. Wood, David Wratt, Panmao Zhai, Tingjun Zhang, De'er Zhang, Xiaoye Zhang, Zong-Ci Zhao, Francis Zwiers*

Union of Concerned Scientists

Brenda Ekwurzel, Peter Frumhoff, Amy Lynd Luers

Channel 4 “The Great Global Warming Swindle” documentary (2007)

Bert Bolin, Piers Corbyn*, Eigil Friis-Christensen, James Shitwaki, Frederick Singer, Carl Wunsch*

Wikipedia’s list of global warming “skeptics”

Khabibullo Ismailovich Abdusamatov*, Syun-Ichi Akasofu*, Sallie Baliunas, Tim Ball, Robert Balling*, Fred Barnes, Joe Barton, Joe Bastardi, David Bellamy, Tom Bethell, Robert Bidinotto, Roy Blunt, Sonja Boehmer, Andrew Bolt, John Brignell*, Nigel Calder, Ian Castles*, George Chilingarian, John Christy*, Ian Clark, Philip Cooney, Robert Davis, David Deming*, David Douglass, Lester Hogan, Craig Idso, Keith Idso, Sherwood Idso, Zbigniew Jaworowski, Wibjorn Karlen, William Kininmonth, Nigel Lawson, Douglas Leahey, David Legates, Richard Lindzen*, Ross Mckittrick*, Patrick Michaels, Lubos Motl*, Kary Mullis, Tad Murty, Tim Patterson, Benny Peiser*, Ian Plimer, Arthur Robinson, Frederick Seitz, Nir Shaviv, Fred Smith, Willie Soon, Thomas Sowell, Roy Spencer, Philip Stott, Hendrik Tennekes, Jan Veizer, Peter Walsh, Edward Wegman

Other sources

Daniel Abbasi, Augie Auer, Bert Bolin, Jonathan Boston, Daniel Botkin*, Reid Bryson, Robert Carter*, Ralph Chapman, Al Gore, Kirtland C. Griffin*, David Henderson, Christopher Landsea*, Bjorn Lomborg, Tim Osborn, Roger Pielke*, Henrik Saxe, Thomas Schelling*, Matthew Sobel, Nicholas Stern*, Brian Valentine*, Carl Wunsch*, Antonio Zichichi.

Appendix B

Forecasting Principles and the Audit Results for Global Warming

The authors made a formal, independent audit of IPCC Chapter 8 in May 2007. Using the Forecasting Audit Software on the forecastingprinciples.com site, ratings were made on a 5-point scale from -2 to +2. A rating of +2 indicates the forecasting procedures were consistent with a principle, and a rating of -2 indicates failure to comply with a principle. Occasionally, some aspects of a procedure are consistent with a principle when others are not. In such cases, the rater must judge where the balance lays. The Audit software also has options to indicate if there is insufficient information to rate the procedures or that the principle is not relevant to a particular forecasting problem.⁴⁴

The audit showed initial overall average ratings were similar at -1.37 and -1.35. The authors compared ratings for each principle and discussed inconsistencies. In some cases ratings were averaged, truncating toward zero. In other cases the authors determined there was insufficient information or that the information was too ambiguous to rate with confidence.⁴⁵

[For the results, see the “Checklist of Forecasting Principles.”]

Checklist of Forecasting Principles from an Audit of the IPCC's *Fourth Assessment Report: Climate Change 2007*

1 = Principles Violated; 2 = Apparent Violations; 3 = Insufficient Information; 4 = N/A

1	2	3	4	Forecasting Principles
PROBLEM				
1. Setting Objectives				
✓				1.1. Describe decisions that might be affected
✓				1.2. Agree on actions for different possible forecasts
✓				1.3. Make forecast independent of organizational politics
✓				1.4. Consider whether events or series are forecastable
	✓			1.5. Gain decision makers' agreement on methods
2. Structuring the Problem				
	✓			2.1. Identify possible outcomes prior to making forecasts
		✓		2.2. Tailor the level of data aggregation to the decisions
		✓		2.3. Decompose the problem into parts
		✓		2.4. Decompose time series by causal forces
		✓		2.5. Structure problems to deal with important interactions
		✓		2.6. Structure problems that involve causal chains
	✓			2.7. Decompose time series by level and trend
INFORMATION				
3. Identifying Information Sources				
		✓		3.1. Use theory to guide information search on explanatory variables
	✓			3.2. Ensure that data match the forecasting situation
✓				3.3. Avoid biased data sources
			✓	3.4. Use diverse sources of data
	✓			3.5. Obtain information from similar (analogous) series or cases
4. Collecting Data				
✓				4.1. Use unbiased and systematic procedures to collect data
✓				4.2. Ensure that information is reliable

1	2	3	4	Forecasting Principles
✓				4.3. Ensure information is valid
		✓		4.4. Obtain all important data
		✓		4.5. Avoid collection of irrelevant data
			✓	4.6. Obtain the most recent data
5. Preparing Data				
		✓		5.1. Clean the data
		✓		5.2. Use transformations as required by expectations
		✓		5.3. Adjust intermittent series
		✓		5.4. Adjust for unsystematic past events (outliers)
		✓		5.5. Adjust for systematic events (e.g., seasonality)
			✓	5.6. Use multiplicative adjustments for seasonality for stable series with trends
			✓	5.7. Damp seasonal factors for uncertainty
		✓		5.8. Use graphical displays for data
METHODS				
6. Selecting Methods				
✓				6.1. Develop list of all important criteria
✓				6.2. Ask unbiased experts to rate potential methods
			✓	6.3. Use structured forecasting methods rather than unstructured
			✓	6.4. Use quantitative methods rather than qualitative methods
			✓	6.5. Use causal rather than naïve methods
✓				6.6. Select simple methods unless evidence favors complex methods
			✓	6.7. Match forecasting method(s) to the situation
✓				6.8. Compare track records of various methods
✓				6.9. Assess acceptability and understandability of methods to users
✓				6.10. Examine value of alternative forecasting methods
7. Implementing Methods: General				
✓				7.1. Keep methods simple
			✓	7.2. Provide a realistic representation of the forecasting situation
✓				7.3. Be conservative in situations of uncertainty or instability
			✓	7.4. Do not forecast cycles

1	2	3	4	Forecasting Principles
		✓		7.5. Adjust for expected events in future
		✓		7.6. Pool similar types of data
		✓		7.7. Ensure consistency with forecasts of related series
8. Implementing Methods: Judgment				
			✓	8.1. Pretest questions used to solicit judgmental forecasts
			✓	8.2. Use questions that have been framed in alternative ways
		✓		8.3. Ask experts to justify their forecasts
			✓	8.4. Use numerical scales with several categories
	✓			8.5. Obtain forecasts from heterogeneous experts
			✓	8.6. Obtain intentions or expectations from representative samples
		✓		8.7. Obtain forecasts from enough respondents
		✓		8.8. Obtain multiple estimates of an event from each expert
9. Implementing Method: Quantitative				
✓				9.1. Tailor the forecasting model to the horizon
		✓		9.2. Match model to underlying process
✓				9.3. Do not use fit to develop a model
		✓		9.4. Weight the most relevant data more heavily
		✓		9.5. Update models frequently
10. Implementing Methods: Quantitative Models with Explanatory Variables				
			✓	10.1. Use theory and domain expertise to select casual variables
		✓		10.2. Use all important variables
		✓		10.3. Use theory and domain expertise to specify directions of relationships
		✓		10.4. Use theory and domain expertise to estimate or limit the magnitude of relationships
		✓		10.5. Use different types of data to estimate a relationship
			✓	10.6. Forecast for at least two alternative environments
		✓		10.7. Forecast for alternative interventions
✓				10.8. Apply the same principles to the forecasts of the explanatory variables
✓				10.9. Shrink the forecasts of change if there is high uncertainty for predictions of the explanatory variables
11. Integrating Judgmental and Quantitative Methods				
✓				11.1. Use structured procedures to do the integration
✓				11.2. Use structured judgment as inputs to models

1	2	3	4	Forecasting Principles
✓				11.3. Use pre-specified domain knowledge as input in selecting, weighting, and modifying quantitative methods
		✓		11.4. Limit subjective adjustments of quantitative forecasts
			✓	11.5. Use judgmental bootstrapping instead of expert forecasts
12. Combining Forecasts				
✓				12.1. Combine forecasts from approaches that differ
			✓	12.2. Use many approaches (or forecasters), preferably at least five
		✓		12.3. Use formal procedures to combine forecasts
		✓		12.4. Start with equal weights
✓				12.5. Use trimmed means, medians, or modes
✓				12.6. Use evidence on each method's accuracy to vary the weights on the component forecasts.
		✓		12.7. Use domain knowledge to vary the weights on the component forecasts
			✓	12.8. Combine when there is uncertainty about which method is best
			✓	12.9. Combine when uncertainty exists about situation
			✓	12.10. Combine when it is important to avoid large errors
EVALUATION				
13. Evaluating Methods				
✓				13.1. Compare reasonable methods
		✓		13.2. Use objective tests of assumptions
	✓			13.3. Design test situation to match the forecasting problem
	✓			13.4. Describe conditions associated with the forecasting problem
✓				13.5. Tailor the analysis to the decision
✓				13.6. Describe potential forecaster biases
✓				13.7. Assess reliability and validity of the data
✓				13.8. Provide easy access to the data
✓				13.9. Provide full disclosure of methods
✓				13.10. Test assumptions for validity
✓				13.11. Test client's understanding of the methods
✓				13.12. Use direct replications of the evaluations to identify mistakes
✓				13.13. Use replications of the forecast evaluations to assess reliability
			✓	13.14. Use extensions of evaluations for generalizability

1	2	3	4	Forecasting Principles
			✓	13.15. Conduct extensions of evaluations in realistic situations
✓				13.16. Compare forecasts generated by different methods
✓				13.17. Examine all important criteria
✓				13.18. Specify criteria prior to analyzing the data
✓				13.19. Assess face validity
✓				13.20. Use error measures that adjust for scale
✓				13.21. Ensure error measures are valid
✓				13.22. Use error measures that are not sensitive to degree of difficulty in forecasting
		✓		13.23. Avoid biased error measure
✓				13.24. Avoid error measures with high sensitivity to outliers
	✓			13.25. Use multiple measures of accuracy
✓				13.26. Use out-of-sample (<i>ex ante</i>) error measures
			✓	13.27. Use <i>ex post</i> accuracy test to evaluate effects
		✓		13.28. Do not use R-square (either standard or adjusted) to compare forecasting models
✓				13.29. Tests of statistical significance should not be used
✓				13.30. Do not use root-mean-square errors to make comparisons
✓				13.31. Base comparisons on large sample
✓				13.32. Conduct explicit cost-benefit analyses
14. Assessing Uncertainty				
			✓	14.1. Estimate prediction intervals (PI)
✓				14.2. Use objective procedures
✓				14.3. Develop prediction intervals by using empirical estimates based on realistic representations of forecasting situations
			✓	14.4. Use transformations when needed to estimate symmetric PIs
		✓		14.5. Ensure consistency over forecast horizon
			✓	14.6. List reasons why forecast might be wrong
✓				14.7. Consider likelihood of alternative outcomes in assessing PIs
✓				14.8. Obtain good feedback on accuracy and reasons for errors
✓				14.9. Combine PIs from alternative forecast methods
✓				14.10. Use safety factors for overconfidence in PIs
			✓	14.11. Conduct experiments
	✓			14.12. Do not assess uncertainty in a traditional group meeting

1	2	3	4	Forecasting Principles
	✓			14.13. Incorporate the uncertainty for predictions of the explanatory variables
		✓		14.14. Ask for a judgmental likelihood that a forecast will fall within a pre-defined minimum-maximum interval
USING FORECASTS				
15. Presenting Forecasts				
✓				15.1. Present forecasts and supporting data in a simple and understandable form
✓				15.2. Provide complete, simple, and clear explanations of methods
	✓			15.3. Describe assumptions
✓				15.4. Present prediction intervals
			✓	15.5. Present forecasts as scenarios
16. Learning				
			✓	16.1. Consider use of adaptive models
		✓		16.2. Seek feedback about forecasts
✓				16.3. Use a formal review process for forecasting methods
✓				16.4. Establish a formal review process to ensure that forecasts are used properly

About the Authors

Kesten C. Green is a senior research fellow of the Business and Economic Forecasting Unit of Monash University. He is codirector of forecastingprinciples.com and is a pioneer of methods to predict the decisions people will make in conflict situations such as occur in wars and in business. Dr. Kesten has published in *Energy and Environment*, *Interfaces*, *Foresight*, *International Journal of Business* and the *International Journal of Forecasting*. He received a doctor of philosophy degree from Victoria University of Wellington.

J. Scott Armstrong is a professor at the Wharton School, University of Pennsylvania. Dr. Armstrong was a founder of the *Journal of Forecasting*, the *International Journal of Forecasting* and the International Symposium on Forecasting. He is the creator of the Web site forecastingprinciples.com and editor of *Principles of Forecasting* (Boston, Mass.: Kluwer, 2001), an evidence-based summary of knowledge on forecasting. In 1996, Armstrong was selected as one of the first six “Honorary Fellows” by the International Institute of Forecasters. He received a doctor of philosophy degree from the Massachusetts Institute of Technology in 1968.

About the NCPA

The NCPA is a nonprofit, nonpartisan organization established in 1983. Its aim is to examine public policies in areas that have a significant impact on the lives of all Americans — retirement, health care, education, taxes, the economy, the environment — and to propose innovative, market-driven solutions. The NCPA seeks to unleash the power of ideas for positive change by identifying, encouraging and aggressively marketing the best scholarly research.

Health Care Policy. The NCPA is probably best known for developing the concept of Health Savings Accounts (HSAs), previously known as Medical Savings Accounts (MSAs). NCPA President John C. Goodman is widely acknowledged (*Wall Street Journal*, *WebMD* and the *National Journal*) as the “Father of HSAs.” NCPA research, public education and briefings for members of Congress and the White House staff helped lead Congress to approve a pilot MSA program for small businesses and the self-employed in 1996 and to vote in 1997 to allow Medicare beneficiaries to have MSAs. In 2003, as part of Medicare reform, Congress and the president made HSAs available to all nonseniors, potentially revolutionizing the entire health care industry. Health Savings Accounts now are potentially available to 250 million nonelderly Americans.

The NCPA outlined the concept of using federal tax credits to encourage private health insurance and helped formulate bipartisan proposals in both the Senate and the House. The NCPA and Blue-Cross Blue-Shield of Texas developed a plan to use money federal, state and local governments now spend on indigent health care to help the poor purchase health insurance. The SPN Medicaid Exchange, an initiative of the NCPA for the State Policy Network, is identifying and sharing the best ideas for health care reform with researchers and policymakers in every state.

Taxes & Economic Growth. The NCPA helped shape the progrowth approach to tax policy during the 1990s. A package of tax cuts designed by the NCPA and the U.S. Chamber of Commerce in 1991 became the core of the Contract with America in 1994. Three of the five proposals (capital gains tax cut, Roth IRA and eliminating the Social Security earnings penalty) became law. A fourth proposal — rolling back the tax on Social Security benefits — passed the House of Representatives in summer 2002. The NCPA’s proposal for an across-the-board tax cut became the centerpiece of President Bush’s tax cut proposals.

NCPA research demonstrates the benefits of shifting the tax burden on work and productive investment to consumption. An NCPA study by Boston University economist Laurence Kotlikoff analyzed three versions of a consumption tax: a flat tax, a value-added tax and a national sales tax. Based on this work, Dr. Goodman wrote a full-page editorial for *Forbes* (“A Kinder, Gentler Flat Tax”) advocating a version of the flat tax that is both progressive and fair.

A major NCPA study, *Wealth, Inheritance and the Estate Tax*, completely undermines the claim by proponents of the estate tax that it prevents the concentration of wealth in the hands of financial dynasties. Actually, the contribution of inheritances to the distribution of wealth in the United States is surprisingly small. Senate Majority Leader Bill Frist (R-TN) and Senator Jon Kyl (R-AZ) distributed a letter to their colleagues about the study. In his letter, Sen. Frist said, “I hope this report will offer you a fresh perspective on the merits of this issue. Now is the time for us to do something about the death tax.”

Retirement Reform. With a grant from the NCPA, economists at Texas A&M University developed a model to evaluate the future of Social Security and Medicare, working under the direction of Thomas R. Saving, who for years was one of two private-sector trustees of Social Security and Medicare.

The NCPA study *Ten Steps to Baby Boomer Retirement* shows that as 77 million baby boomers begin to retire, the nation’s institutions are totally unprepared. Promises made under Social Security, Medicare and Medicaid are completely unfunded. Private sector institutions are not doing better — millions of workers are discovering

that their defined benefit pensions are unfunded and that employers are retrenching on post-retirement health care promises.

Pension reforms signed into law include ideas to improve 401(k)s developed and proposed by the NCPA and the Brookings Institution. Among the NCPA/Brookings 401(k) reforms are automatic enrollment of employees into the companies' 401(k) plans, automatic contribution rate increases so that as workers' wages grow so do their contributions, and stronger default investment options for workers who do not make an investment choice.

The NCPA's online Social Security calculator allows visitors to discover their expected taxes and benefits and how much they would have accumulated had their taxes been invested privately.

Environment & Energy. The NCPA's E-Team is one of the largest collections of energy and environmental policy experts and scientists who believe that sound science, economic prosperity and protecting the environment are compatible. The team seeks to correct misinformation and promote sensible solutions to energy and environment problems. A pathbreaking 2001 NCPA study showed that the costs of the Kyoto agreement to reduce carbon emissions in developed countries would far exceed any benefits.

Educating the next generation. The NCPA's Debate Central is the most comprehensive online site for free information for 400,000 U.S. high school debaters. In 2006, the site drew more than one million hits per month. Debate Central received the prestigious Templeton Freedom Prize for Student Outreach.

Promoting Ideas. NCPA studies, ideas and experts are quoted frequently in news stories nationwide. Columns written by NCPA scholars appear regularly in national publications such as the *Wall Street Journal*, the *Washington Times*, *USA Today* and many other major-market daily newspapers, as well as on radio talk shows, on television public affairs programs, and in public policy newsletters. According to media figures from Burrelle's, more than 900,000 people daily read or hear about NCPA ideas and activities somewhere in the United States.

What Others Say About the NCPA

"The NCPA generates more analysis per dollar than any think tank in the country. It does an amazingly good job of going out and finding the right things and talking about them in intelligent ways." –Newt Gingrich, former Speaker of the U.S. House of Representatives

"We know what works. It's what the NCPA talks about: limited government, economic freedom; things like health savings accounts. These things work, allowing people choices. We've seen how this created America." –John Stossel, co-anchor ABC-TV's 20/20

"I don't know of any organization in America that produces better ideas with less money than the NCPA." –Phil Gramm, former U.S. Senator

"Thank you . . . for advocating such radical causes as balanced budgets, limited government and tax reform, and to be able to try and bring power back to the people." –Tommy Thompson, former Secretary of Health and Human Services

The NCPA is a 501(c)(3) nonprofit public policy organization. We depend entirely on the financial support of individuals, corporations and foundations that believe in private sector solutions to public policy problems. You can contribute to our effort by mailing your donation to our Dallas headquarters or logging on to our Web site at www.ncpa.org and clicking "An Invitation to Support Us."